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# Price Optimization

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# Introduction

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- Choice model captures demand
- Demand is elastic to price
- Predicted demand varies with price, if it is a variable of the model
- In principle, the probability to use/purchase an alternative decreases if the price increases.
- The revenue per user increases if the price increases.
- Question: what is the optimal price to optimize revenue?

In short:

- Price $\uparrow$  $\Rightarrow$  profit/passenger $\uparrow$  and number of passengers  $\downarrow$
- Price $\downarrow$  $\Rightarrow$  profit/passenger $\downarrow$  and number of passengers  $\uparrow$
- What is the best trade-off?

# Revenue calculation

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Number of persons choosing alternative  $i$  in the population

$$\hat{N}(i) = \sum_{s=1}^S N_s P(i|x_s, p_{is})$$

where

- $p_s$  is the price of item  $i$  in segment  $s$
- $x_s$  gathers all other variables corresponding to segment  $s$
- the population is segmented into  $S$  homogeneous strata
- $P(i|x_s, p_{is})$  is the choice model
- $N_s$  is the number of individuals in segment  $s$

# Revenue calculation

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The total revenue from  $i$  is therefore:

$$R_i = \sum_{s=1}^S N_s P(i|x_s, p_{is}) p_{is}$$

If the price is constant across segments, we have

$$R_i = p_i \sum_{s=1}^S N_s P(i|x_s, p_i)$$

# Price optimization

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Optimizing the price of product  $i$  is solving the problem

$$\max_{p_i} p_i \sum_{s=1}^S N_s P(i|x_s, p_i)$$

Notes:

- It assumes that everything else is equal
- In practice, it is likely that the competition will also adjust the prices

# Illustrative example

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A binary logit model with

$$V_1 = \beta_p p_1 - 0.5$$

$$V_2 = \beta_p p_2$$

so that

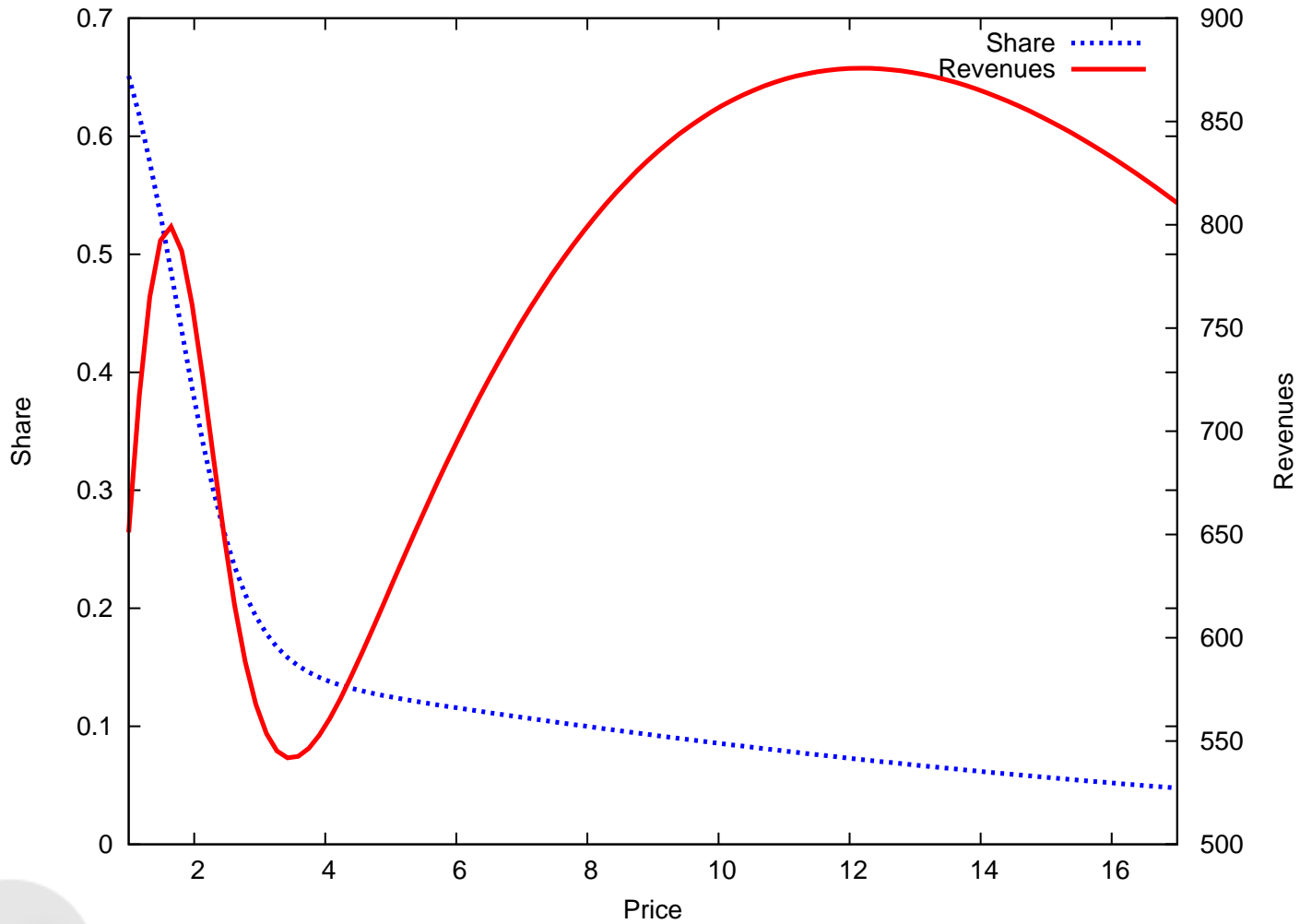
$$P(1|p) = \frac{e^{\beta_p p_1 - 0.5}}{e^{\beta_p p_1 - 0.5} + e^{\beta_p p_2}}$$

Two groups in the population:

- Group 1:  $\beta_p = -2$ ,  $N_s = 600$
- Group 2:  $\beta_p = -0.1$ ,  $N_s = 400$

Assume that  $p_2 = 2$ .

# Illustrative example



# Sensitivity analysis

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- Parameters are estimated, we do not know the real value
- 95% confidence interval:  $[\hat{\beta}_p - 1.96\sigma, \hat{\beta}_p + 1.96\sigma]$
- Perform a sensitivity analysis for  $\beta_p$  in group 2



# Sensitivity analysis

